

Predictive Toxicology and In Vitro to In Vivo Extrapolation

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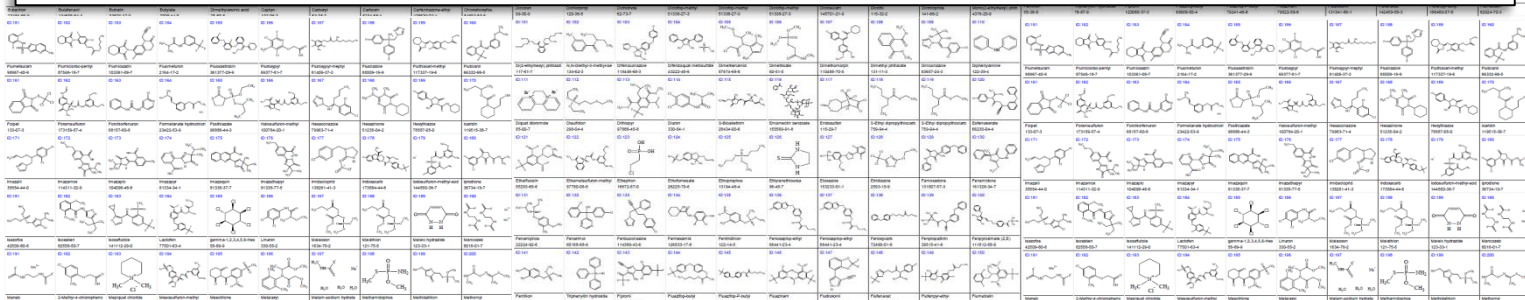


AsiaTox 2015

25 June 2015, Jeju City Korea

Too many chemicals to test with standard animal-based methods

- Cost, time, animal welfare



Need for better mechanistic data

- Determine human relevance
- What is the Adverse Outcome Pathway (AOP)?

-
- The figure is a schematic representation of the proposed model, divided into three main panels. The left panel shows a 'Lactococcal cell' with 'Monoglycerols' and 'Fatty acids' being released from a 'Micelle'. The middle panel shows a 'Macrophage' with 'Lipid droplet' and 'Vesicle' containing 'VLDL' and 'LDL'. The right panel shows a 'Macrophage cell' with 'CD36, CXCR4, IL-1, TNF-α', 'KCNK', 'Vesicle remodeling', and 'Cell membrane' components. The bottom panel shows a 'KEY' with symbols for 'Established non-cholesterol link age cells', 'Predictable link age cells', 'Emphasis of link age based on quantitative response data', 'Predictive model link age based on quantitative response data', 'Macrophage link age', and 'Activity link age'.

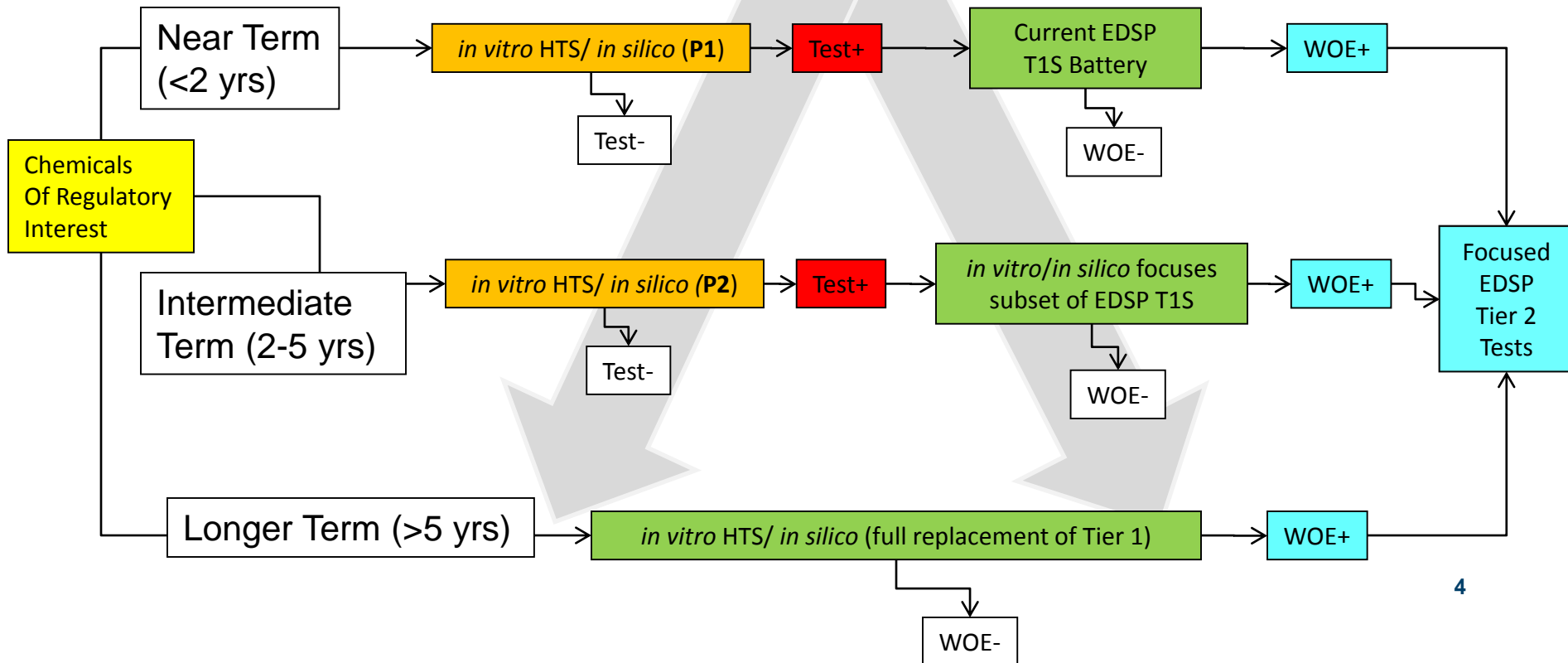
Computational Toxicology

- Identify biological pathways of toxicity (AOPs)
- Develop high-throughput *in vitro* assays to test chemicals
- Test “Human Exposure Universe” chemicals in the assays
- Develop models that link *in vitro* to *in vivo* hazard
- Use pharmacokinetic models to predict activating doses
- Develop exposure models for all chemicals

CompTox and the Endocrine Disruptor Screening Program

- 10,000 chemicals to be tested
- 100-200 years, \$Billions of cost with current tests
- Need methods to prioritize chemicals
- Need high-throughput, lower cost replacement tests

EPA Research provides basis for improving the suite of assays and models to advance chemical prioritization and screening



In Vitro Estrogen Receptor Model

Combines results from multiple in vitro assays

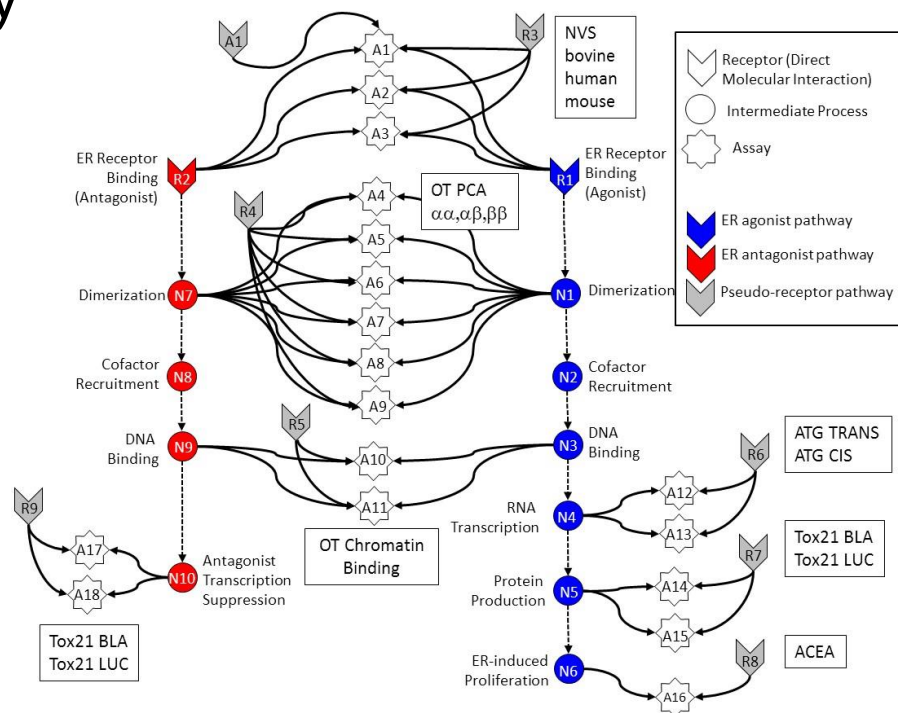
- Use multiple assays per pathway
 - Different technologies
 - Different points in pathway

- No assay is perfect
 - Assay Interference
 - Noise

- Use model to integrate assays

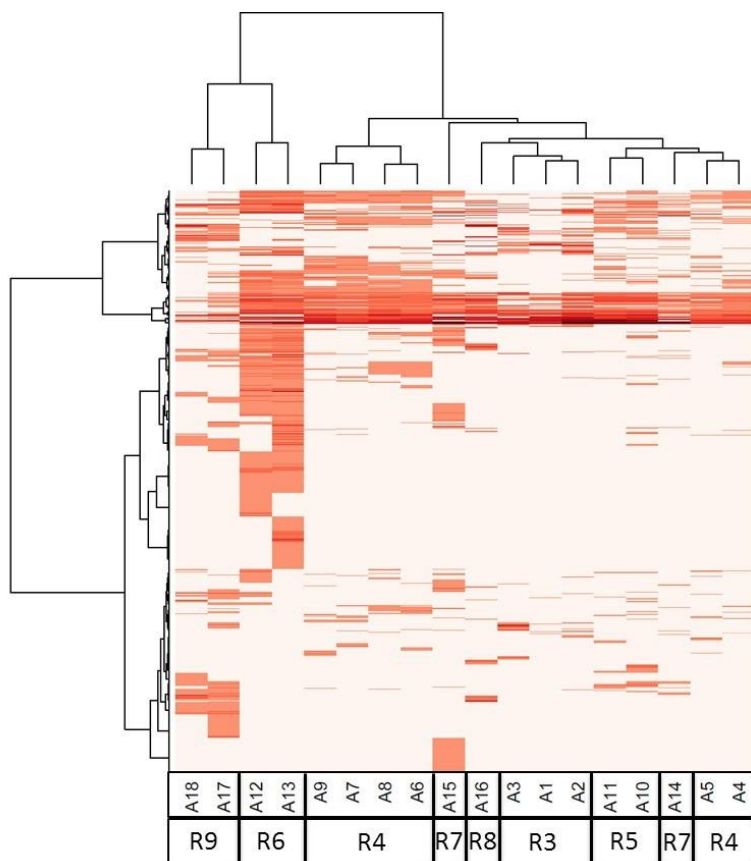
- Evaluate model against reference chemicals

- Methodology being applied to other pathways



Major theme – all assays have false positives and negative

Assays cluster by technology, suggesting technology-specific non-ER bioactivity

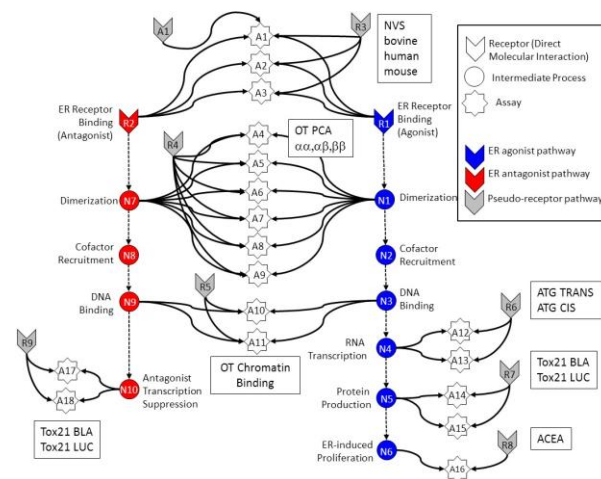


Much of this “noise” is reproducible

- “assay interference”
- Result of interaction of chemical with complex biology in the assay

EDSP chemical universe is structurally diverse

- Solvents
- Surfactants
- Intentionally cytotoxic compounds
- Metals
- Inorganics
- Pesticides
- Drugs

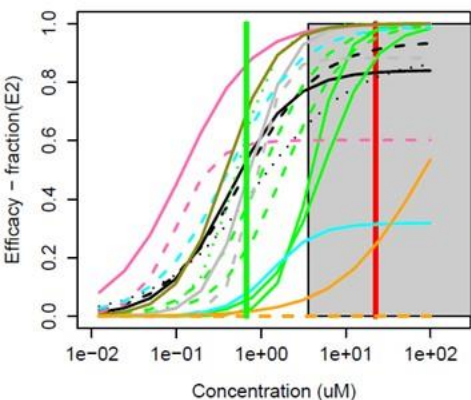


Example curves

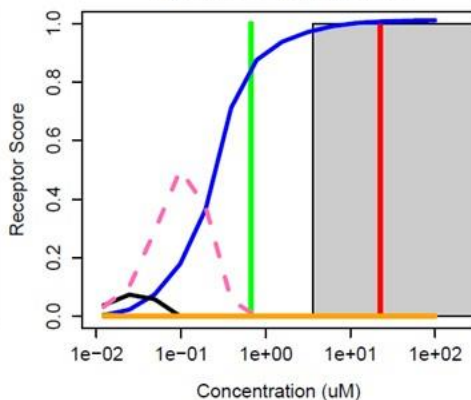
True Agonist

True Antagonist

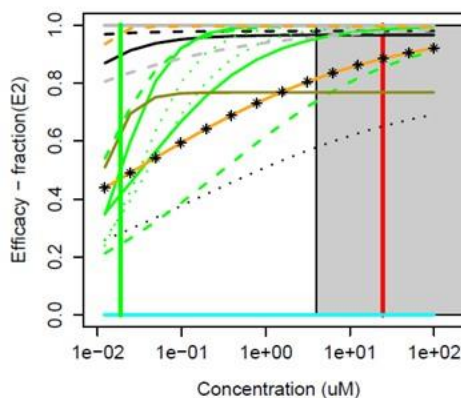
80-05-7 : Bisphenol A



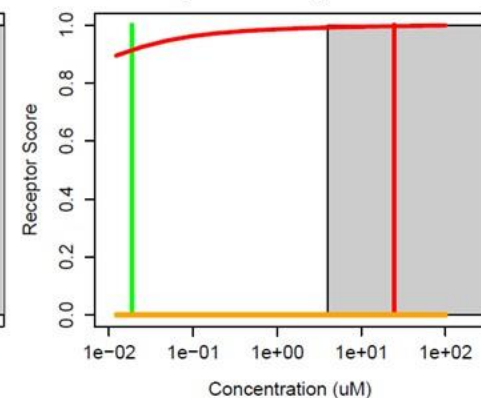
80-05-7 : Bisphenol A
Agonist: 0.65 Antagonist: 0



82640-04-8 : Raloxifene hydrochloride

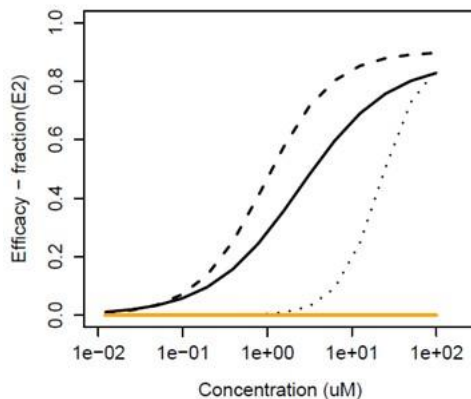


82640-04-8 : Raloxifene hydrochloride
Agonist: 0 Antagonist: 0.97

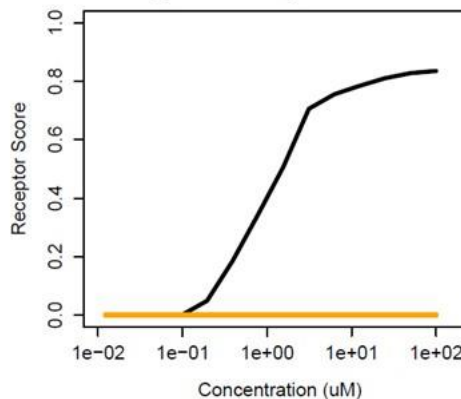


Negative-Narrow Assay Interference

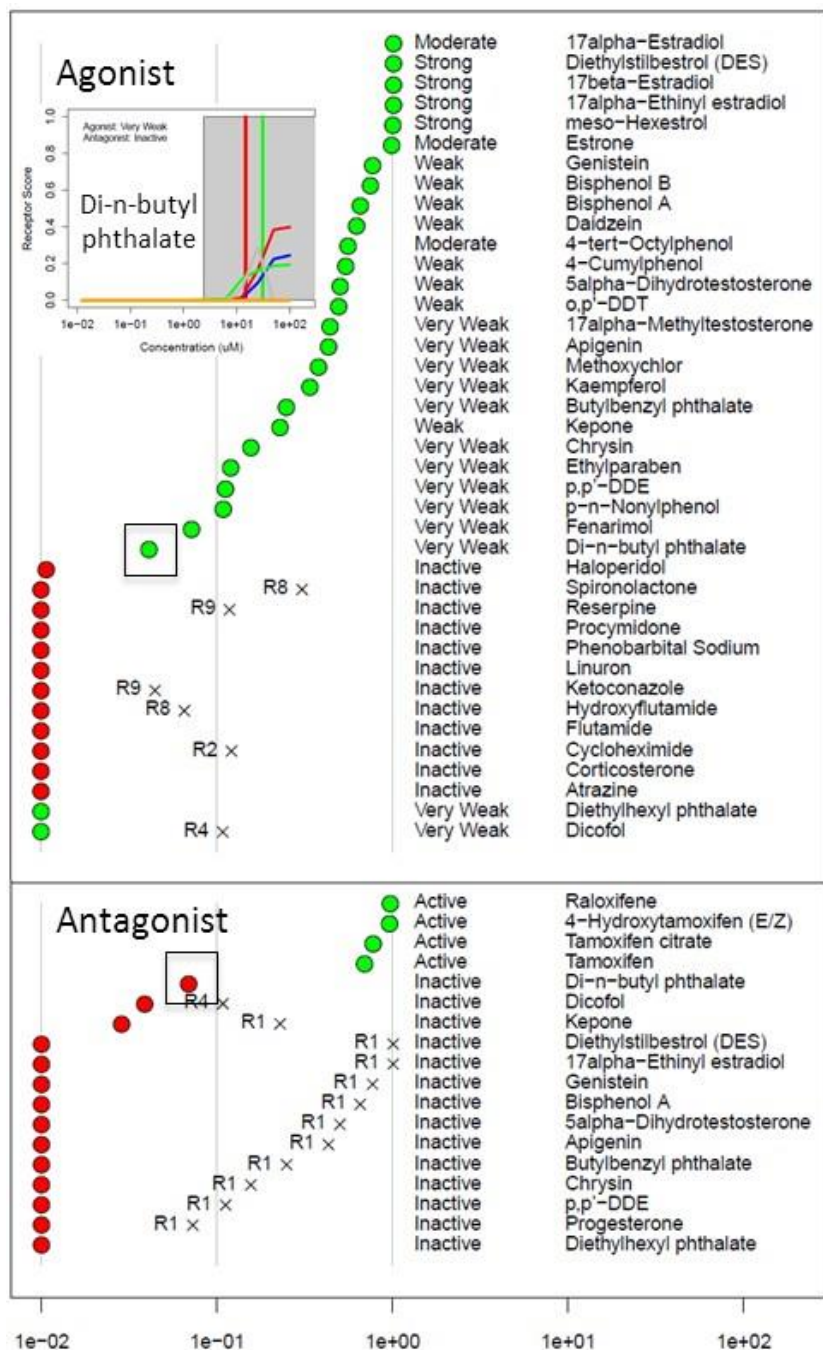
10016-20-3 : alpha-Cyclodextrin



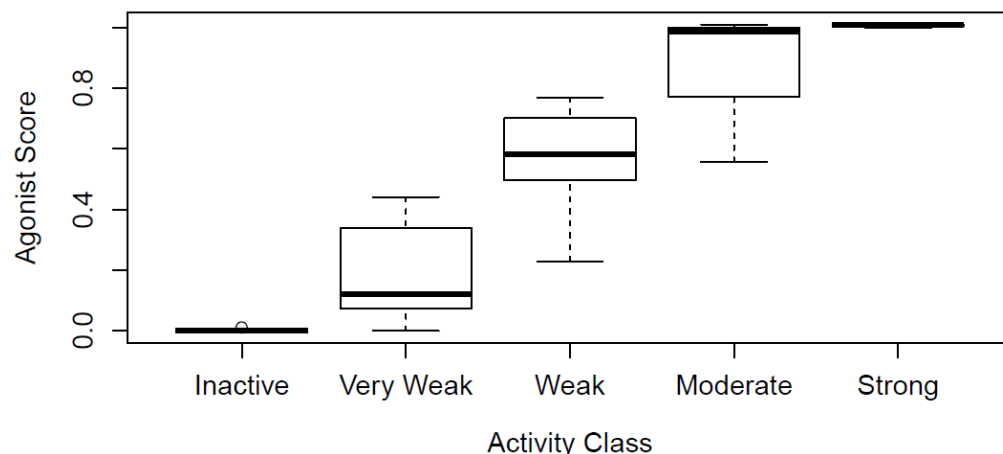
10016-20-3 : alpha-Cyclodextrin
Agonist: 0 Antagonist: 0.00022



In Vitro Reference Chemical Performance



Agonist Score (R1) vs. Reference Activity Class

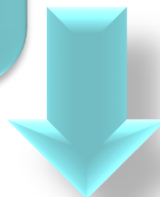


***In Vivo* Reference Chemicals: Guideline Uterotrophic Assay Data**

Uterotrophic Literature
“Guideline-Like” Studies
(start with 700 papers)

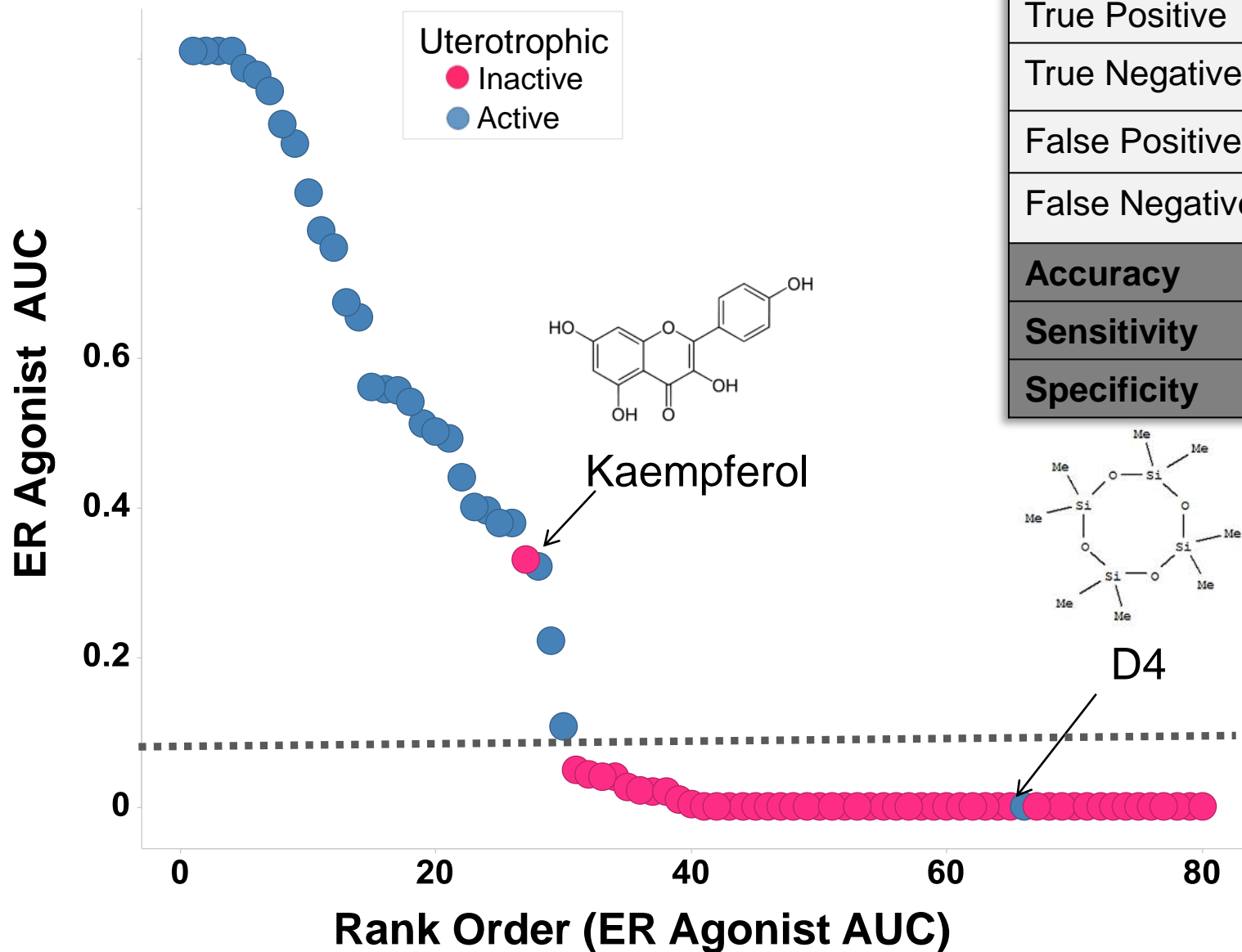
+

EDSP List 1 Uterotrophic
“Guideline” Studies



Uterotrophic Reference Chemicals:
30 Active, 51 Inactive

ER Agonist AUC vs. Uterotrophic Outcomes

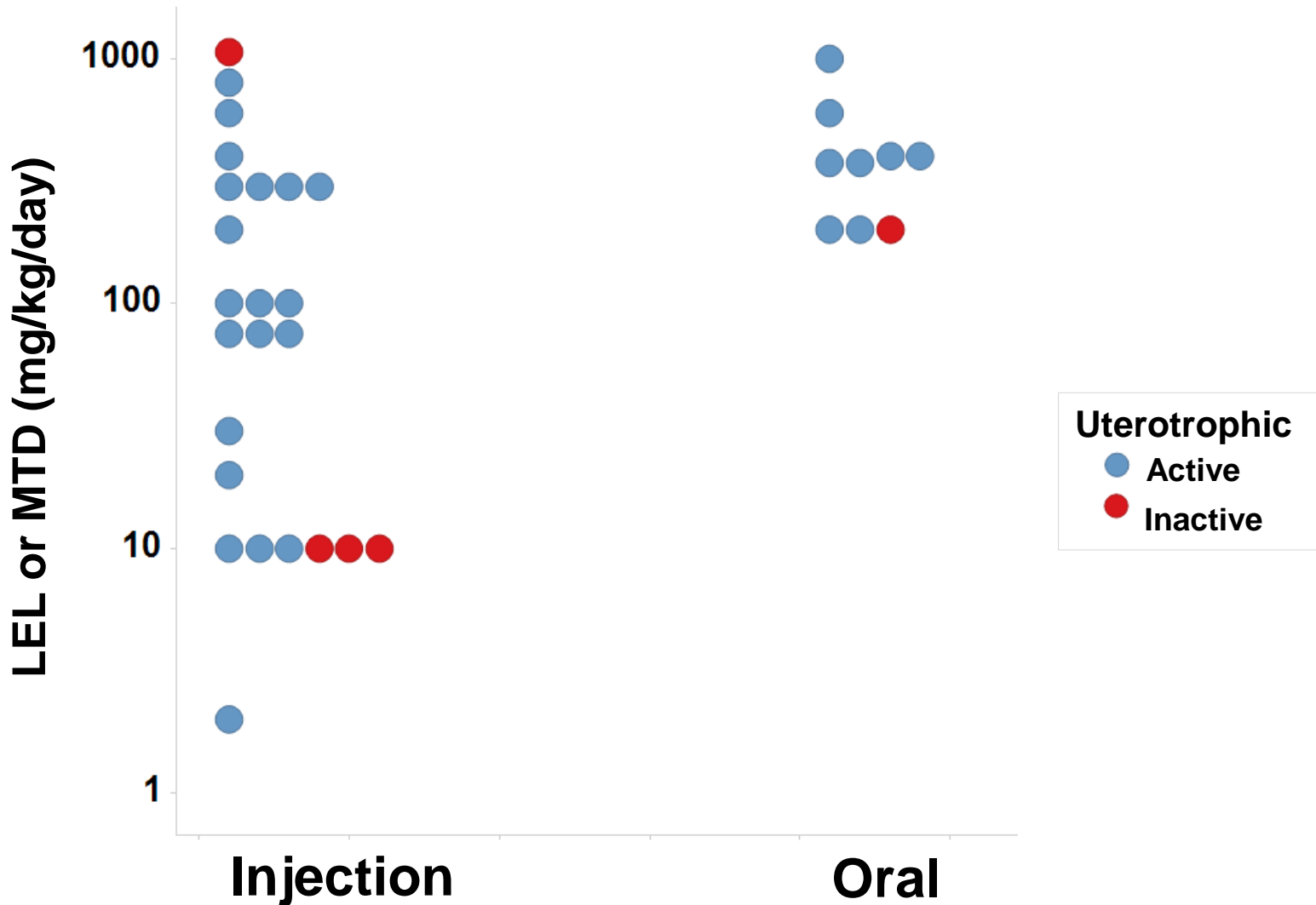


True Positive	29
True Negative	50
False Positive	1
False Negative	1
Accuracy	0.97
Sensitivity	0.97
Specificity	0.98

Browne et al. Screening chemicals for estrogen receptor bioactivity using a computational model (ES&T in press)

In vivo* guideline studies have the same types of uncertainty as *in vitro

Immature Rat: BPA



- Goal: To make ER and AR data easily available to all stakeholders
 - Assay-by-assays concentration-response plots
 - Model scores – AUC agonist and antagonist
 - ER QSAR calls
 - Other relevant data
- <http://actor.epa.gov/edsp21>

The screenshot shows the EDSP21 Dashboard interface. On the left, there's a 'Chemical Selection' sidebar with filters for CASRN, Chemical Name, and IsToxC. The main area displays 'Chemical Structure and Data' for Bisphenol A. It includes a chemical structure diagram, a table of properties (DISSTOX GSID, CASRN, Name, SMILES, InChI, InChI Key, Molecular Wt, Chemical Formula, Cytotoxicity Limit, Chemical Type, Chiral/Stereo, dbt/Stereo, Organic Form, Iupac), and a 'PhysChem Properties' table with columns for Property, Model Name, Raw Result, Result (Mean), Result (min), Result (max), and Result Unit. The dashboard also has tabs for Chemical Summary, Public Information, Bioactivity Summary, Bioactivity, High-Throughput Exposure, Assay Definitions, and Dashboard.

ToxCast Model Predictions		
Model	Agonist AUC	Antagonist AUC
ER	0.45	0
AR	0	0.136

Consensus CERAPP QSAR ER Model Predictions			
Class	Agonist (Potency Level)	Antagonist (Potency Level)	Binding (Potency Level)
from Literature	Active (Weak)	-	Active (Weak)
QSAR Consensus	Active (Weak)	Active (Strong)	Active (Weak)

Moving Towards Regulatory Acceptance From FIFRA SAP, December 2014

- Can the ER Model be used for prioritization?
 - “... the ER AUC appears to be an **appropriate tool for chemical prioritization** for ... the EDSP universe compounds.”
- Can the ER model substitute for the Tier 1 ER in vitro and uterotrophic assays?
 - “... **replacement of the Tier 1 *in vitro* ER endpoints ...with the ER AUC model will likely be a more effective and sensitive measure for the occurrence of estrogenic activity...**”
 - “... the Panel **did not recommend that the uterotrophic assay be substituted** by the AUC model at this time. The Panel suggested that the EPA considers: 1) conducting limited uterotrophic and other Tier 1 in vivo assay testing, using the original Tier 1 Guidelines (and/or through literature curation)”
- Based on follow-up presented here (FR notice, June 18 2015) ...
 - “**EPA concludes that ER Model data are sufficient to satisfy the Tier 1 ER binding, ERTA and uterotrophic assay requirements.**”

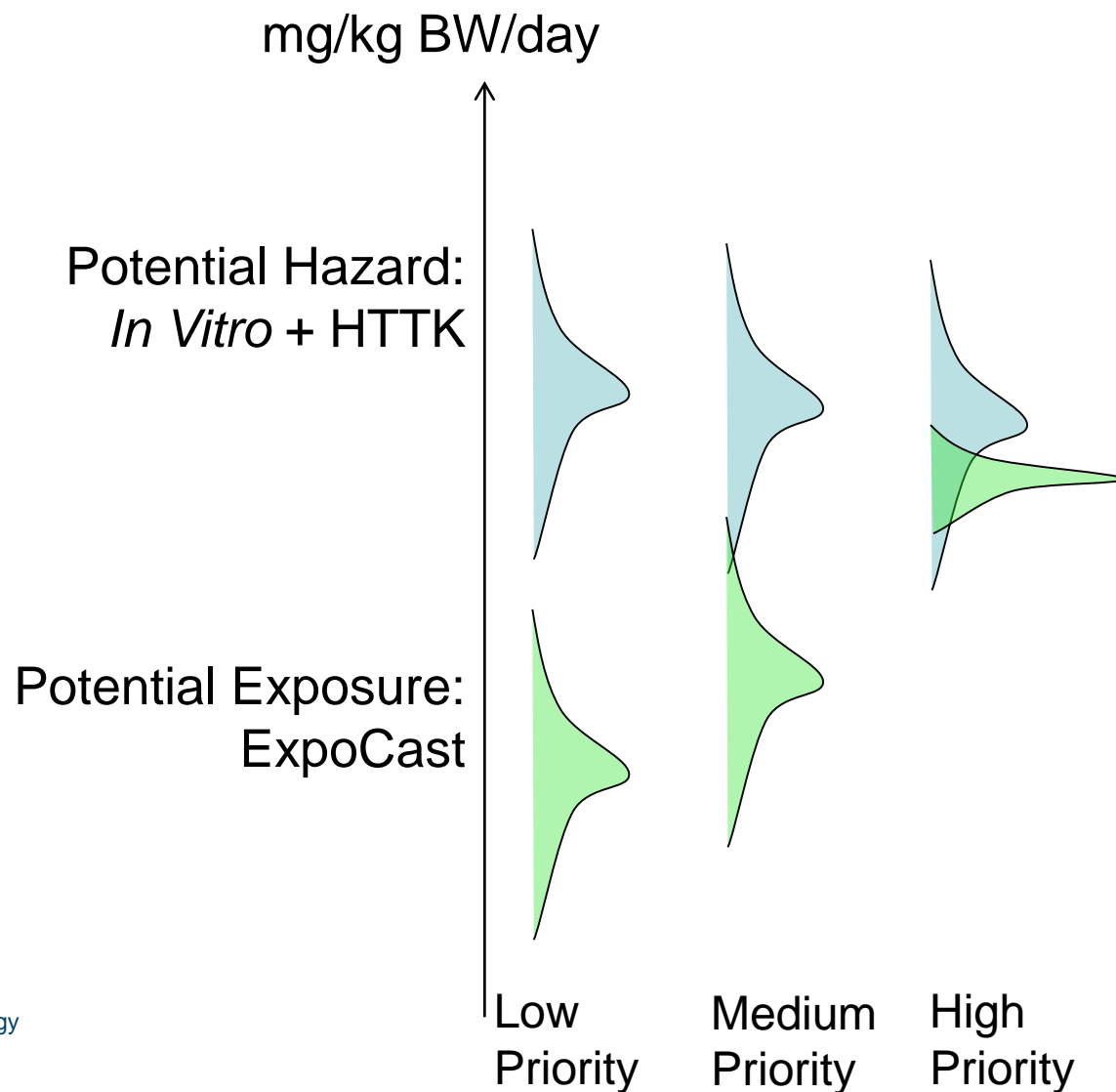
Modeling Thyroid Disruption

- Develop assays for key targets
 - Thyroid hormone receptor (Complete)
 - Thyroid peroxidase (TPO) (Screening in progress)
 - Deiodinases (assays in development)
 - NIS – Sodium-Iodide Symporter (assays in development)
 - Transporters (assays planned)
- Screen Chemicals
- Predict *in vivo* potency for assay hits
- Test effects in complex “tissue on a chip” systems

Risk-based Prioritization

Hazard + Exposure

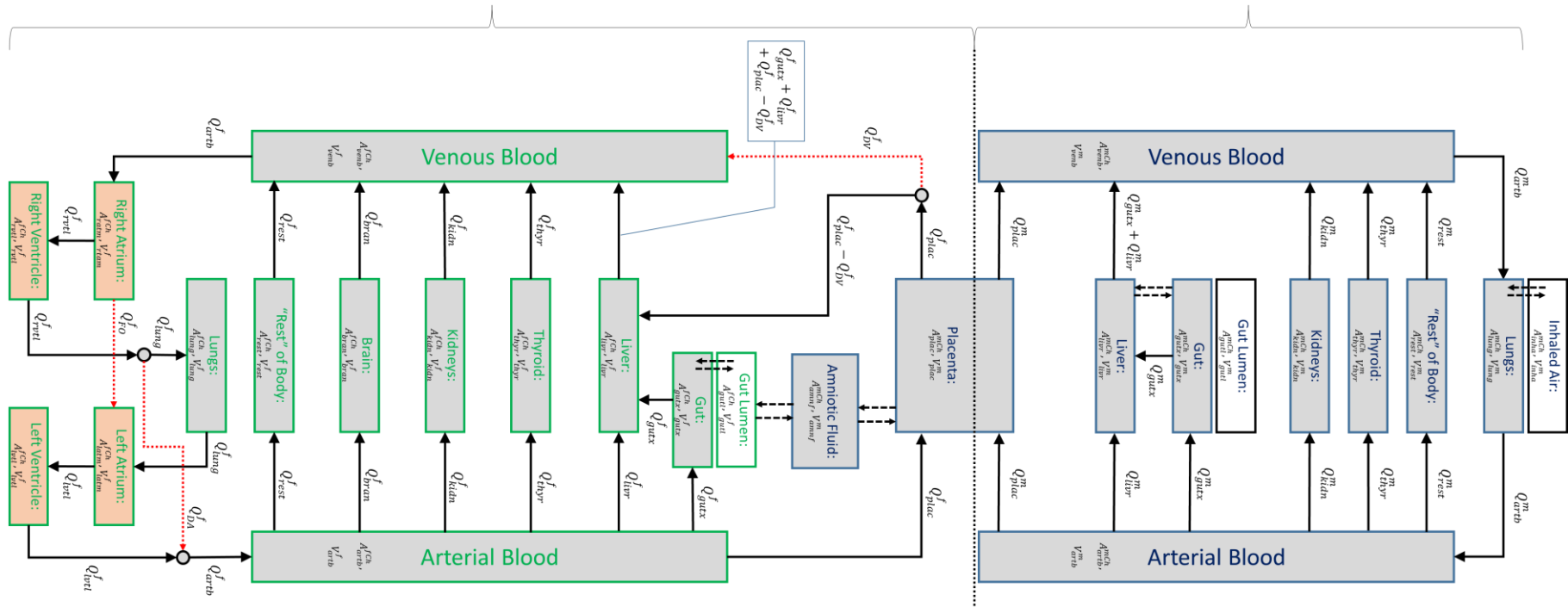
Semi-quantitative
In Vitro to *In Vivo*
Approach



Maternal/Fetal PBPK Model

Fetal Blood

Maternal Blood

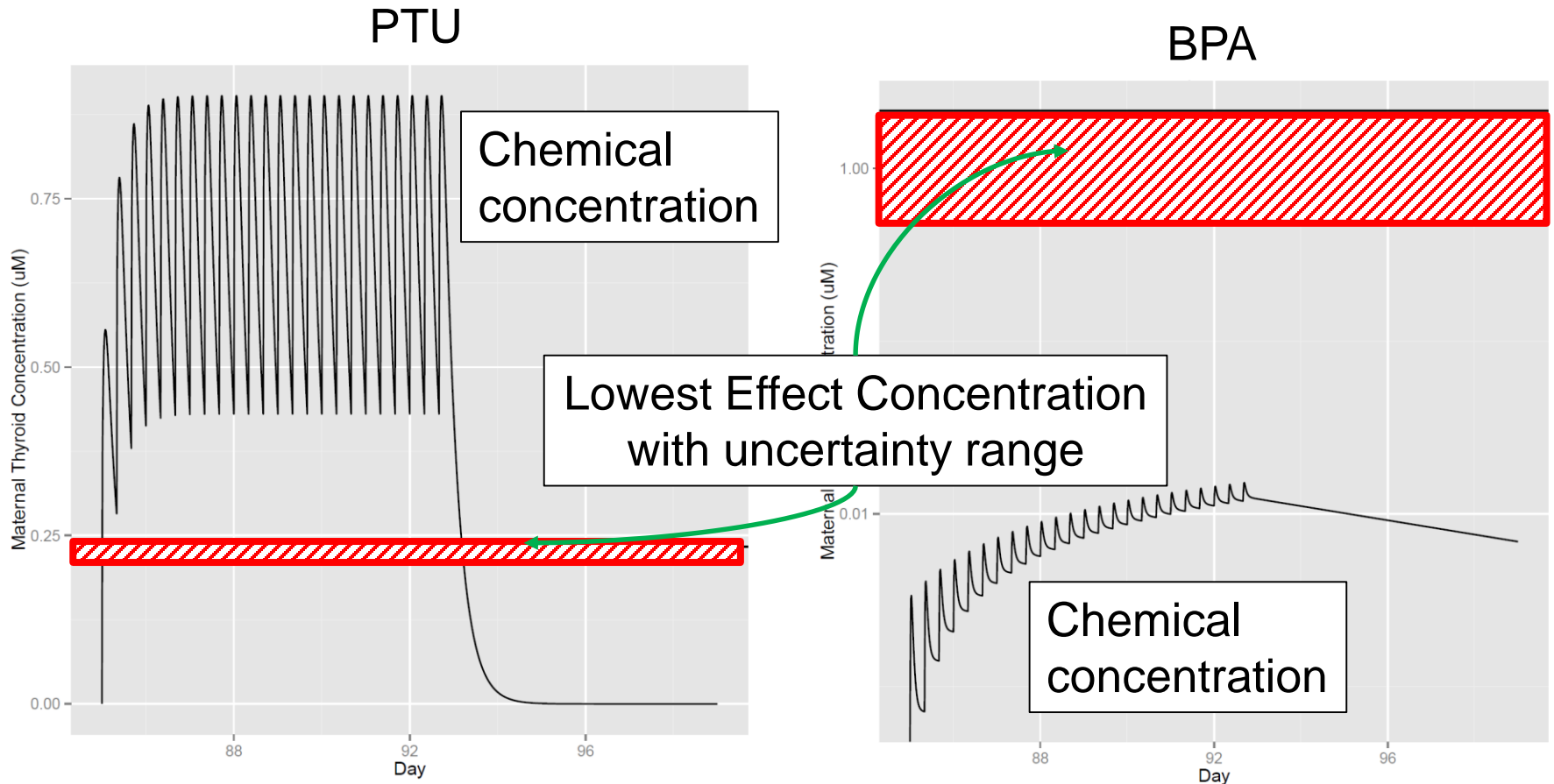


Model accounts for development of fetus, weeks 12 to term

Parameters for ~500 chemicals

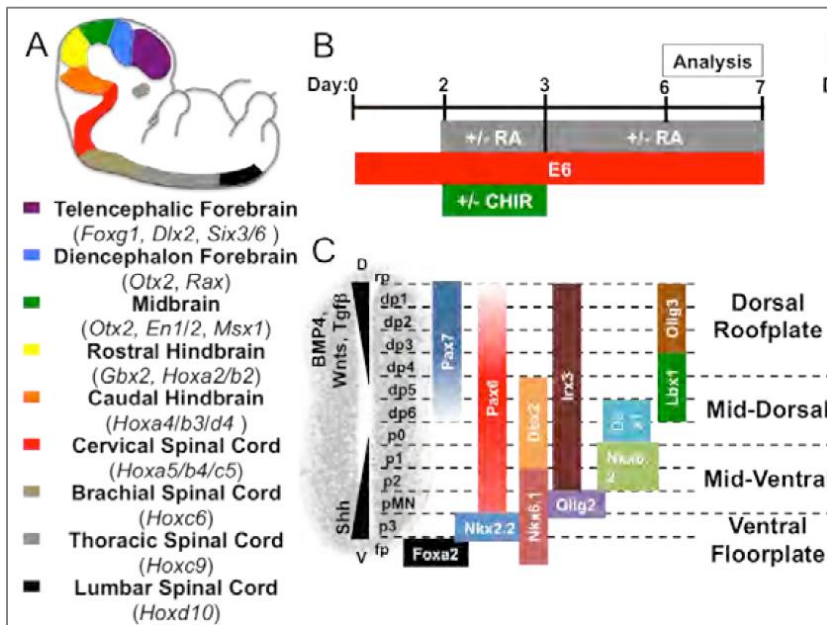
Prioritizing Chemicals Using the PBPK Model

PTU and BPA both target TPO (thyroid peroxidase)
But ... effect of 1 mg/kg/day is much different

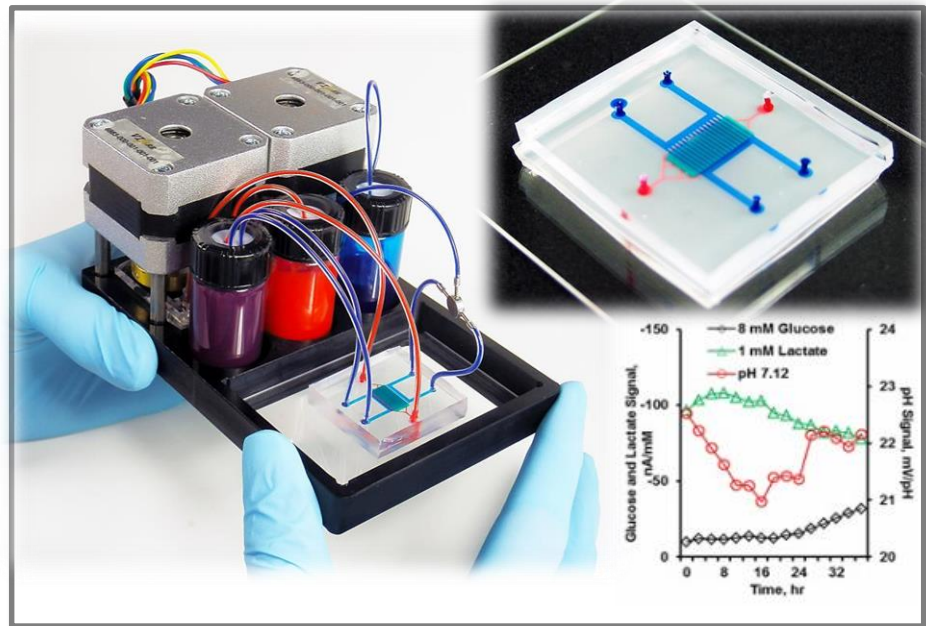


Brain Development and the Neurovascular Unit

Randy Ashton (UWisc): 3D hCNS microsystem derived from hPSCs and patterned for phenotypic diversity across 9 discrete body axis domains.



John Wikswo (Vanderbilt): synthetic BBB (endothelia/pericyte/astrocyte) channel interfaced via porous matrix to neuron/microglia/WBC channel.



Collaboration to a synthetic model for thyrotropic neurodevelopment

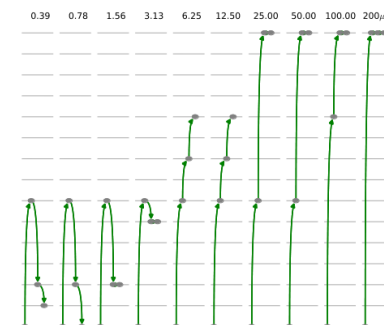
Key Strengths and Weaknesses of *In Vitro* Systems for Toxicity Testing

- Strengths

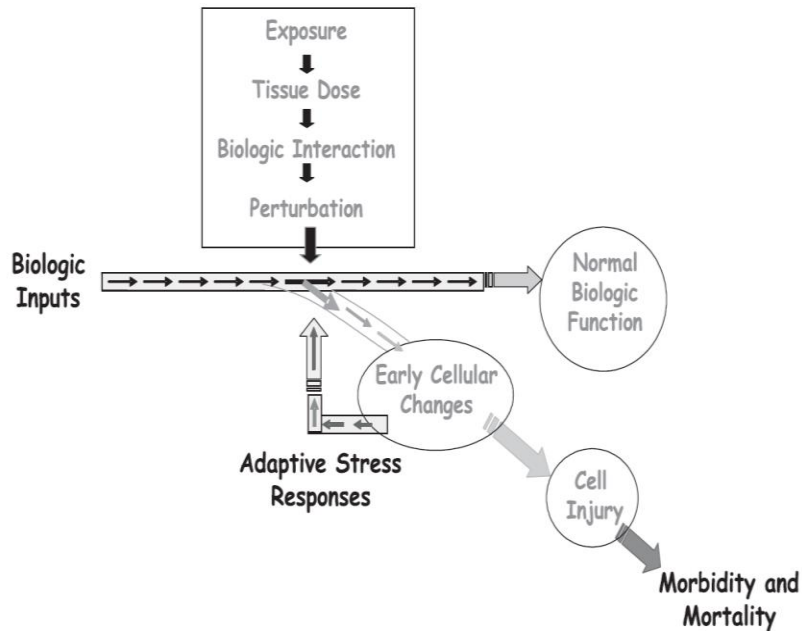
- Rapid development of new assays
- Ability to screen thousands of chemicals
- Direct link to molecular basis of adversity

- Weaknesses

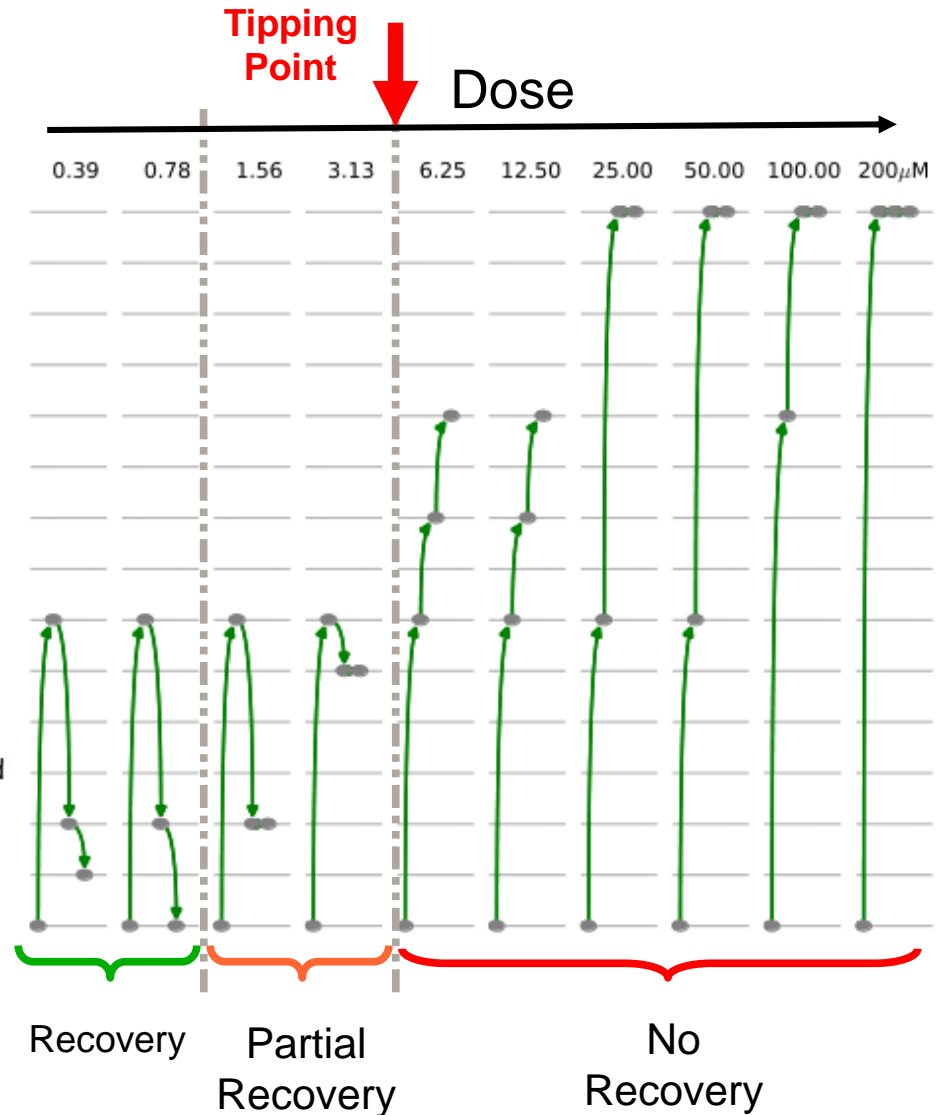
- Often lack metabolic capacity
- Often lack complex key multi-cell type signaling
- Often lack ability to adapt



In Vitro Adaptation “Tipping Points” Use Time-Dose Trajectories



From NAS
Toxicity Testing in the 21st Century



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